```
!unzip '/content/spam.csv'
Archive: /content/spam.csv
  End-of-central-directory signature not found. Either this file is
  a zipfile, or it constitutes one disk of a multi-part archive.
the
  latter case the central directory and zipfile comment will be found
on
  the last disk(s) of this archive.
        cannot find zipfile directory in one of /content/spam.csv or
        /content/spam.csv.zip, and cannot find /content/spam.csv.ZIP,
period.
#Import required library
 import pandas as pd
 import numpy as np
 from sklearn.model selection import train test split
 from sklearn.preprocessing import LabelEncoder
 from keras.models import Model
 from keras.layers import LSTM, Activation, Dense, Dropout, Input,
Embedding
 from keras.optimizers import RMSprop
 from keras.preprocessing.text import Tokenizer
 from keras preprocessing import sequence
 from keras.utils import to categorical
 from tensorflow.keras.models import Sequential
#Read Dataset and do preprocessing
df = pd.read csv('/content/spam.csv',delimiter=',',encoding='latin-
1')
df.head()
     v1
                                                         v2 Unnamed: 2
\
         Go until jurong point, crazy.. Available only ...
                                                                   NaN
1
    ham
                             Ok lar... Joking wif u oni...
                                                                   NaN
        Free entry in 2 a wkly comp to win FA Cup fina...
2
   spam
                                                                   NaN
3
    ham
         U dun say so early hor... U c already then say...
                                                                   NaN
4
    ham Nah I don't think he goes to usf, he lives aro...
                                                                   NaN
```

#Download the dataset

```
Unnamed: 3 Unnamed: 4
0
         NaN
                    NaN
1
         NaN
                    NaN
2
         NaN
                    NaN
3
         NaN
                    NaN
         NaN
                    NaN
 df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed:
4'],axis=1,inplace=True) #dropping unwanted
 df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 2 columns):
     Column Non-Null Count Dtype
 #
 0
     v1
             5572 non-null
                              obiect
 1
     v2
             5572 non-null
                              object
dtypes: object(2)
memory usage: 87.2+ KB
df.groupby(['v1']).size() # Count of Spam and Ham values,
v1
ham
        4825
         747
spam
dtype: int64
# Label Encoding target column
X = df.v2
Y = df.v1
le = LabelEncoder()
Y = le.fit transform(Y)
Y = Y.reshape(-1,1)
 X train, X test, Y train, Y test = train test split(X, Y, test size=0.15)
# Test and train split
 \max \text{ words} = 1000
 \max len = 150
 tok = Tokenizer(num words=max words)
 tok.fit on texts(X train)
 sequences = tok.texts_to_sequences(X_train)
 sequences matrix = sequence.pad sequences(sequences,maxlen=max len)
Create Model and Add Layers (LSTM, Dense-(Hidden Layers), Output)
 #Create Model
 input = Input(name='InputLayer', shape=[max len])
```

## #Add Layers (LSTM, Dense-(Hidden Layers), Output)

```
layer = Embedding(max_words,50,input_length=max_len)(inputs)
layer = LSTM(64)(layer)
layer = Dense(256,name='FullyConnectedLayer1')(layer)
layer = Activation('relu')(layer)
layer = Dropout(0.5)(layer)
layer = Dense(1,name='OutputLayer')(layer)
layer = Activation('sigmoid')(layer)

#Compile The Model

model = Model(inputs=inputs,outputs=layer)
model.summary()
model.compile(loss='binary_crossentropy',optimizer=RMSprop(),metrics=['accuracy'])
```

Model: "model\_1"

Layer (type)	Output Shape	Param #
InputLayer (InputLayer)	[(None, 150)]	0
<pre>embedding_3 (Embedding)</pre>	(None, 150, 50)	50000
lstm_3 (LSTM)	(None, 64)	29440
<pre>FullyConnectedLayer1 (Dense )</pre>	(None, 256)	16640
<pre>activation_6 (Activation)</pre>	(None, 256)	0
<pre>dropout_3 (Dropout)</pre>	(None, 256)	0
OutputLayer (Dense)	(None, 1)	257
activation_7 (Activation)	(None, 1)	0

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Total params: 96,337 Trainable params: 96,337 Non-trainable params: 0

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#Fit The Model

```
Epoch 1/30
- accuracy: 0.8788 - val loss: 0.5301 - val accuracy: 0.8481
Epoch 2/30
- accuracy: 0.8820 - val_loss: 0.5733 - val_accuracy: 0.8354
Epoch 3/30
- accuracy: 0.8881 - val loss: 0.5753 - val accuracy: 0.8302
Epoch 4/30
- accuracy: 0.8928 - val loss: 0.6359 - val accuracy: 0.7806
Epoch 5/30
- accuracy: 0.9010 - val loss: 0.6116 - val accuracy: 0.8175
Epoch 6/30
- accuracy: 0.9055 - val_loss: 0.6498 - val_accuracy: 0.8049
Epoch 7/30
- accuracy: 0.9118 - val loss: 0.6866 - val accuracy: 0.7806
Epoch 8/30
30/30 [============= ] - 7s 234ms/step - loss: 0.2049
- accuracy: 0.9139 - val loss: 0.7650 - val accuracy: 0.8133
Epoch 9/30
- accuracy: 0.9213 - val_loss: 0.7879 - val_accuracy: 0.7711
Epoch 10/30
- accuracy: 0.9242 - val_loss: 0.8410 - val_accuracy: 0.8080
Epoch 11/30
- accuracy: 0.9277 - val loss: 0.8711 - val accuracy: 0.8049
Epoch 12/30
30/30 [============== ] - 7s 238ms/step - loss: 0.1710
- accuracy: 0.9359 - val loss: 0.9013 - val accuracy: 0.7901
Epoch 13/30
30/30 [============== ] - 7s 237ms/step - loss: 0.1683
- accuracy: 0.9348 - val loss: 0.9538 - val accuracy: 0.8070
Epoch 14/30
- accuracy: 0.9382 - val loss: 0.9887 - val accuracy: 0.7943
Epoch 15/30
- accuracy: 0.9435 - val loss: 1.0721 - val accuracy: 0.7932
Epoch 16/30
- accuracy: 0.9424 - val loss: 1.0436 - val accuracy: 0.7932
Epoch 17/30
```

```
- accuracy: 0.9440 - val loss: 1.0637 - val accuracy: 0.7648
Epoch 18/30
30/30 [============= ] - 7s 236ms/step - loss: 0.1404
- accuracy: 0.9475 - val loss: 1.0544 - val accuracy: 0.7985
Epoch 19/30
- accuracy: 0.9514 - val loss: 1.1296 - val accuracy: 0.7584
Epoch 20/30
- accuracy: 0.9520 - val loss: 1.2484 - val accuracy: 0.7711
Epoch 21/30
- accuracy: 0.9535 - val loss: 1.2196 - val accuracy: 0.7679
Epoch 22/30
- accuracy: 0.9591 - val loss: 1.2910 - val accuracy: 0.7426
Epoch 23/30
- accuracy: 0.9575 - val loss: 1.3979 - val accuracy: 0.7542
Epoch 24/30
- accuracy: 0.9591 - val loss: 1.4036 - val accuracy: 0.7637
Epoch 25/30
- accuracy: 0.9622 - val loss: 1.4102 - val accuracy: 0.7806
Epoch 26/30
- accuracy: 0.9617 - val loss: 1.4230 - val_accuracy: 0.7447
Epoch 27/30
30/30 [============= ] - 7s 239ms/step - loss: 0.0929
- accuracy: 0.9644 - val loss: 1.5604 - val accuracy: 0.7110
Epoch 28/30
- accuracy: 0.9654 - val loss: 1.5457 - val accuracy: 0.7743
Epoch 29/30
- accuracy: 0.9649 - val loss: 1.5456 - val accuracy: 0.7437
Epoch 30/30
30/30 [============== ] - 7s 236ms/step - loss: 0.0831
- accuracy: 0.9681 - val loss: 1.7871 - val accuracy: 0.7489
<keras.callbacks.History at 0x7f59d6db4f50>
#SAVE MODEL
model.save('model .h5')
#TEST The MODEL
```

```
test_sequences = tok.texts_to_sequences(X_test)
test sequences matrix =
sequence.pad_sequences(test_sequences,maxlen=max_len)
accuracy = model.evaluate(test sequences matrix,Y test)
print('Accuracy: {:0.3f}'.format(accuracy[1]))
accuracy: 0.7500
Accuracy: 0.750
y_pred = model.predict(test_sequences_matrix)
\overline{\text{print}}(y \text{ pred}[25:40].\text{round}(3))
27/27 [========] - 1s 19ms/step
[[0.]
 [0.998]
 [0.]
 [0.
 [0.002]
 [0.
 [0.
 [0.
 [0.
 [0.681]
 [0.001]
 [0.
 [0.
      ]
 [0.
      1
      ]]
 [0.
```